Politecnico di Milano – Scuola di Ingegneria Industriale e dell'Informazione Academic Year 2020/2021 - FIRST semester Course code 052499 - BAYESIAN STATISTICS - 10 ECTS credits Laurea Magistrale di ING MATEMATICA Master of Science in Mathematical Engineering - LEONARDO Campus

Instructor

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List of the Proofs

In order to help students to pass the written exam, here it is a list of the proofs which you could be asked to prove:

- 1 normalizing constants and moments of the following distributions: gamma, Weibull, beta, Dirichlet.
- properties of the finite-dimentional Dirichlet distribution (13/08/14) (6/7/16)
- ★ Bayes' Theorem for dominated models
- * posterior mean as the functional minimizing the posterior quadratic loss function (multivariate case), posterior median as the functional minimizing the posterior absolute value loss function
- ✗ use of the MC standard error to bound (in probability) the Monte Carlo error
- ✓ definition of the method of composition and its use in Bayesian Statistics
- **★•** Generalized inverse distribution function sampling method:
 - (i) sampling from a truncated distribution function on a real interval, i.e. conditioning on $X \in (a, b)$;

(23/01/18)

- (ii) sampling from a gamma(m, b) distribution, when m is an integer.
- 💢 simulation from a univariate (Box-Müller) and multivariate Gaussian distribution
- acceptance-rejection method to sample from a (univariate) target density f; application: sampling from the gamma (α, β) distribution, with $\alpha > 1$, when the proposal is the gamma(m, b) density with m integer; parameters choice in order to maximize the algorithm efficiency
- Optimal choice for the proposal distribution of the importance sampler, as the one minimizing the variance of the estimator.
- \times reversibility of the Metropolis-Hastings algorithm wrt the target distribution π (14/01/2070)
- Gaussian linear model with homoscedastic (iid) errors: conjugate prior when the variance is known and when the variance is unknown, posteriors and parameter updates; Jeffreys's prior (proportional to $1/\sigma^2$, where σ^2 is the error variance) and its posterior
- formal approach to model choice: calculation of the posterior probability of model j, j = 1, ..., K

examp of Question 2.



• computation of LPML (log-pseudo marginal likelihood) from a MCMC sample from the posterior, given all the datapoints (13/07/2017) (1/3/16)

X • probit regression model with latent variables: computation of the full-conditionals

(03/07/2020)

• likelihood for right-censored (conditionally independent) data (3/2/17)

(05/07/2017)

first and second moments of the Dirichlet process P (included the covariance between P(A) and P(B))

(*) joint marginal distribution of a sample from a Dirichlet process; its interpretation as the generalized

• (*) Pólya urn (with two colors): the sequence of r.v.s representing the color of the sampled balls is exchangeable with beta distribution as de Finetti measure

(*) predictive distribution under the DPM model, marginalizing (i.e. integrating out) the r.p.m. P

(Dirichlet process) (H12/20(6) X • (*) computation of full-conditionals from the Pólya-urn scheme in DPM models (Escobar-West, 1994)

(*) cluster estimates (definition)

💓 • (*) calculation of the optimal partition minimizing a posteriori Binder's loss function via MCMC draws

• all definitions (e.g. Bayes factor, WAIC, ...)

Topics with (*) pertain to the 10 credits course, and do not pertain to the 8 credit version of the course.

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- Bayesian Hierarchical model?
- Beta-Bernouli model (p. 16)
- Bayesian point estimation (p. 17)
- · Bayerian intervals (p. 17)
- Bayerian Hp. testing (p.18)
- Bayenau prediction (p.18) (p.28)
- Normal-normal model (p. 13)
- Exchangeability (p.21) (+ Definetti (p. 22))
- conjugate distributions (p.24)
- Fetfreys' prior (p.25)
- Normal-inverse-gamma (p. 27)
- Metropolis-Hastings (p.39)
- Gibbs toumpler (p. 40)
- Full conditionals (p.53)
- Goodness of fit (p. 70)
- GLM (p.72)

- Gamma (& Inverse Gamma)
- Beta
- Dinchlet
- Bernouli
- Bironnial
- Mutinouial
- Weibull
- + Poidson
- + Exponential
- + Normal
- Unitorm
- t student
- Wishart
- 2 mult. Normal
- 7 mult. t-student

- Poisson-gamme model (8/10/2020) Normal-inverse-mishort (15/10/2020)